

Urban Vulnerability Assessment for the Health and Infrastructure Sectors

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Mission and Year 1 Activities

Explore regional vulnerabilities to climate variability and change in urban areas using an end-to-end approach that combines:

- 1) Downscaled Climate Projections (from CMIP phases 3 and 5 global climate models from GISS and the wider ensemble)
- 2) Urban Heat Island Analysis and Mitigation Scenarios
- 3) Monitoring of Urban Climate and Green Infrastructure

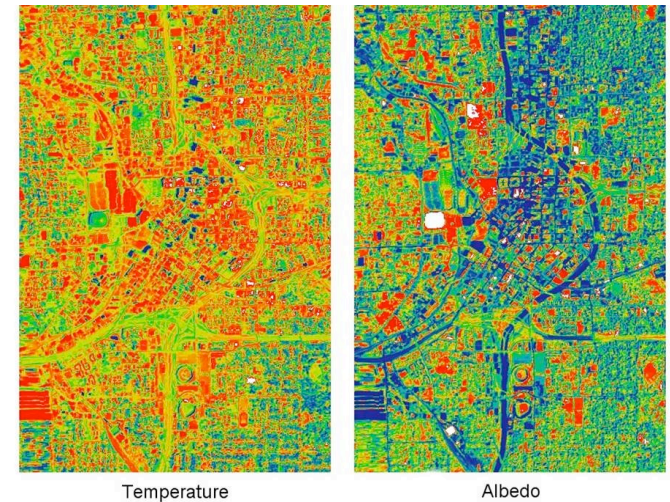


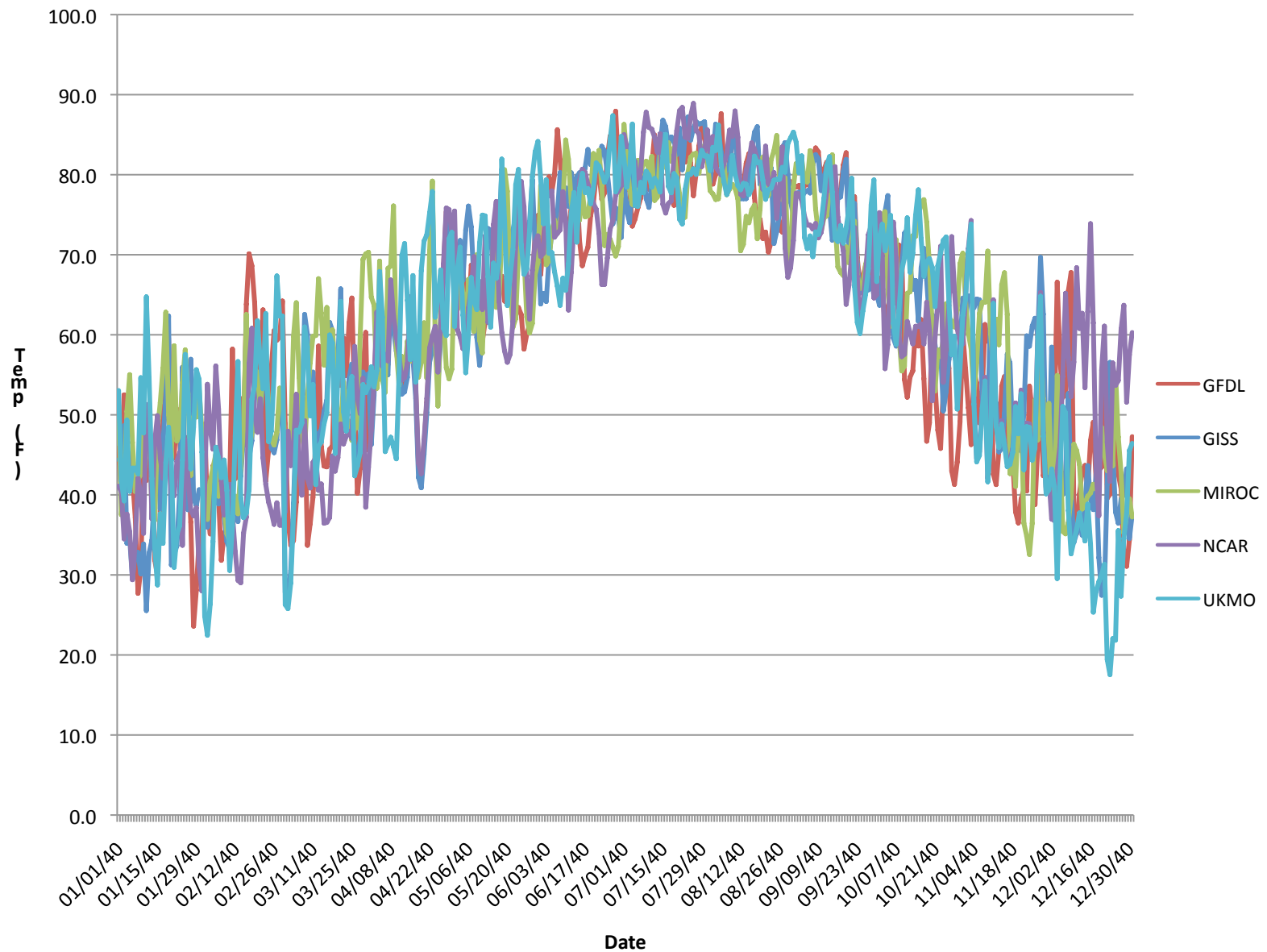
Figure 1. The image on the left illustrates daytime surface heating for urban surfaces across the Atlanta Georgia Central Business District (CBD) as derived from NASA aircraft data. White and red colors indicate very warm surfaces (~40-50°C). Green relates to surfaces of moderately warm temperatures (~25-30°C). Blue indicates cool surfaces (e.g., vegetation, shadows) (~15-20°C). Surface temperatures are reflected in the albedo image on the right where warm surfaces are dark (i.e., low reflectivity) and cooler surfaces are in red and green (i.e., higher reflectivity). The images show how urban surface characteristics influence temperature and albedo as UHI drivers (Quattrochi et al., 2000).

Statistical Downscaling Procedure

- **Datasets required**
 - Observed daily climate data (maximum and minimum temperature) for a single station for a thirty year period (ideally 1970-1999).
 - Future projection climate outputs (monthly) are needed for the gridbox that covers the observed climate station. The outputs are from the 16 GCMs used in the Bias Corrected Spatially Disaggregated (BCSD; Maurer et al. 2007 http://gdo-dcp.ucllnl.org/downscaled_cmip3_projections/dcpInterface.html#Welcome) projections, at 1/8th degree spatial resolution for a given emissions scenario and 30-year timeslice.
- **Step 1** – Compute monthly mean temperature (from daily values) and rank the values separately for each of the 12 calendar months.
- **Step 2** – For each month and year, calculate the daily temperature anomaly (for maximum and minimum) by subtracting the monthly mean from the daily value.
- **Step 3** – Rank the future mean temperature values for each calendar month (for a given scenario, timeslice, and each GCM).
- **Step 4** – Match the ‘rank’ of the first month of the first year of the future series to the same ‘rank’ in the observations. The daily temperature anomalies from the observed month are then added to the future monthly means to form the future daily projected series.
 - For example, if January 2040 (the starting point for the 2050s timeslice) is the 4th warmest January in the 30 year future timeslice, the 4th warmest January in the observed 30 year record is found.
- **Step 5** – Repeat for all 12 months and 30 years for each GCM, timeslice, and emissions scenario.

Downscaling Results

Projected Daily Average Temperatures for Hampton, VA (2040)



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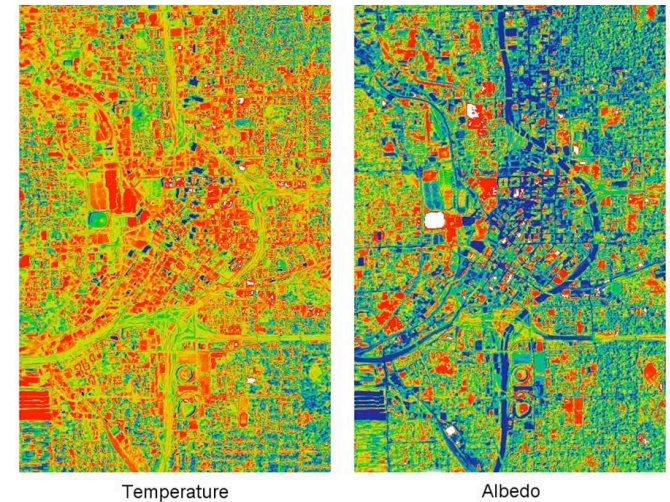
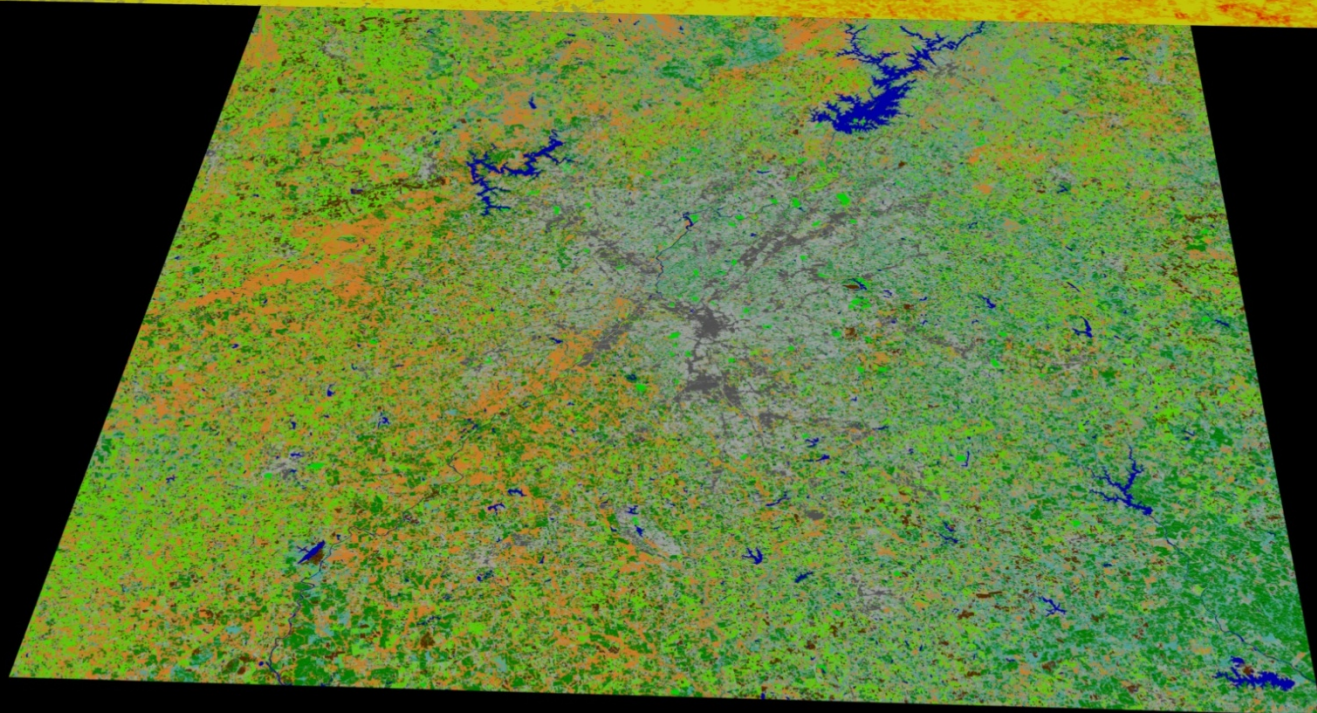
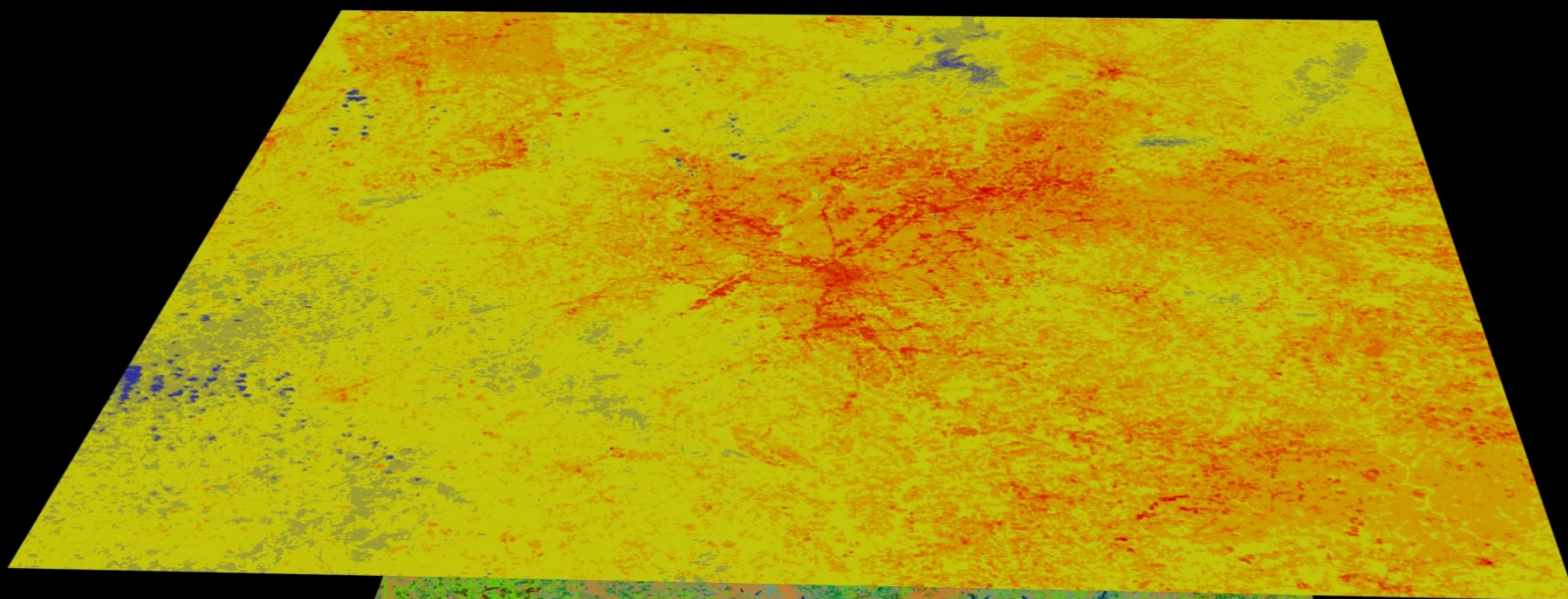
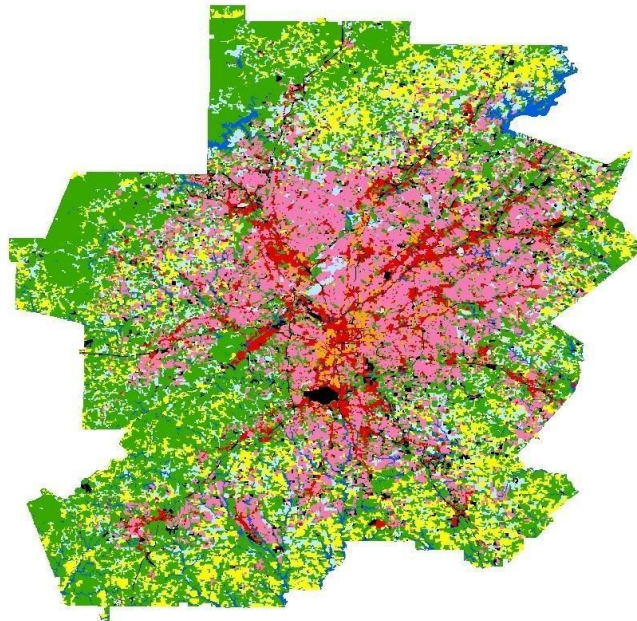


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Current and Projected 2030 Land Use 13-county Atlanta Metro Area

Current (1999)



Projected (2030)

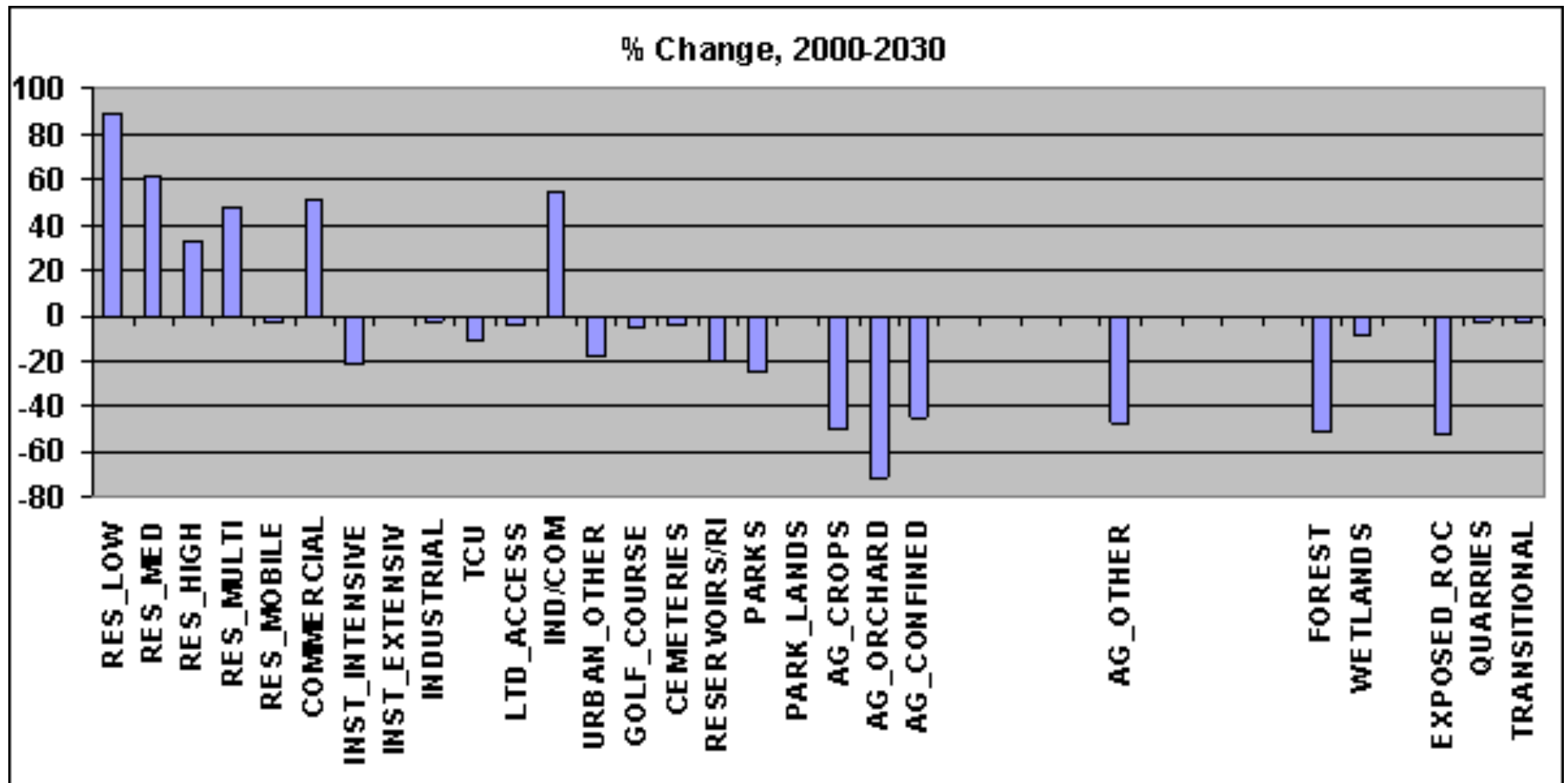


The Spatial Growth Model (SGM) was used to project land use/land cover for the area to 2030.

Inputs to the model are current land use and current and projected population, employment, and road networks.

Current land use/land cover is defined by the LandPro99 data set created by the Atlanta Regional Commission (ARC).

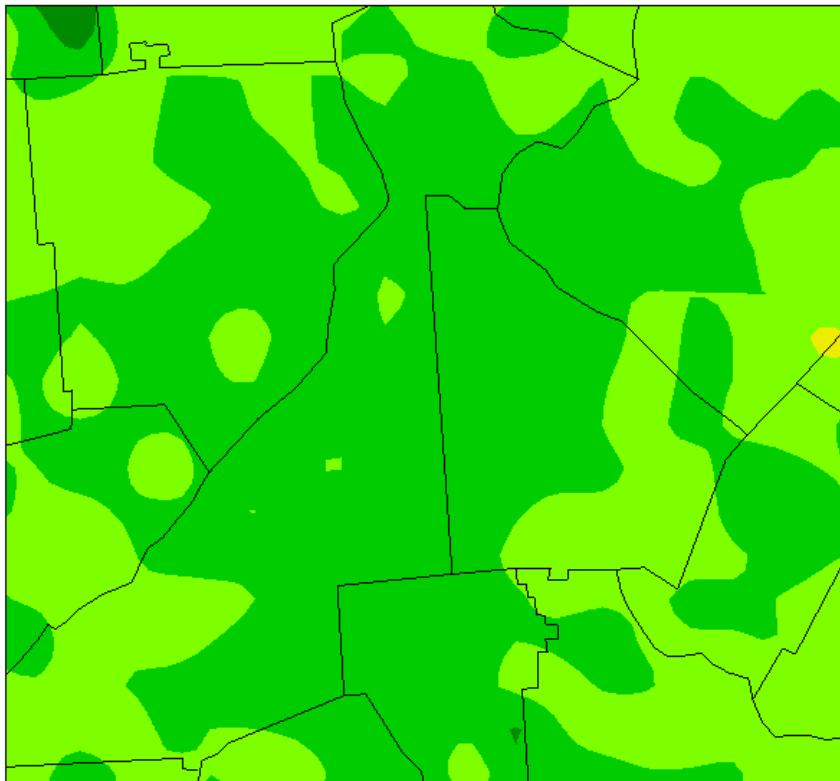
Land Use Projections



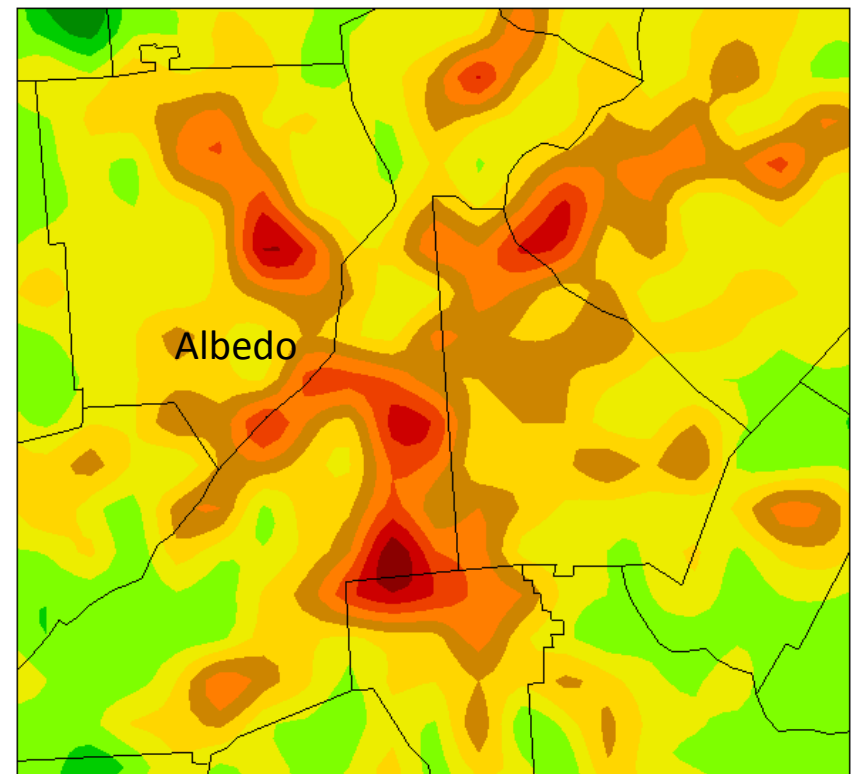
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Albedo Mitigation Simulation - 2030

2030 Business As Usual



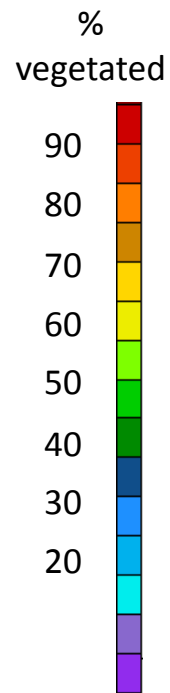
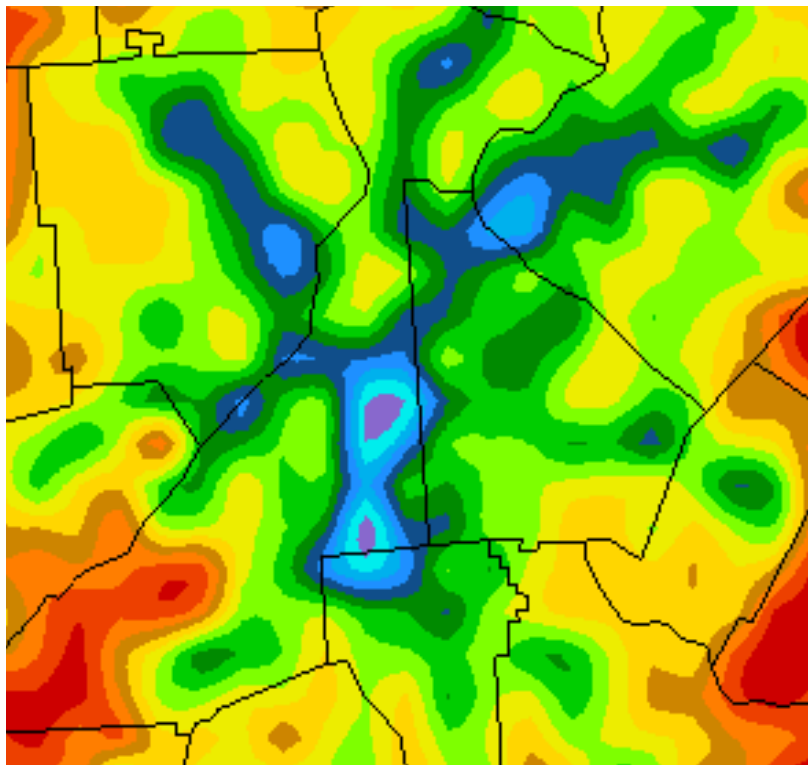
High Albedo



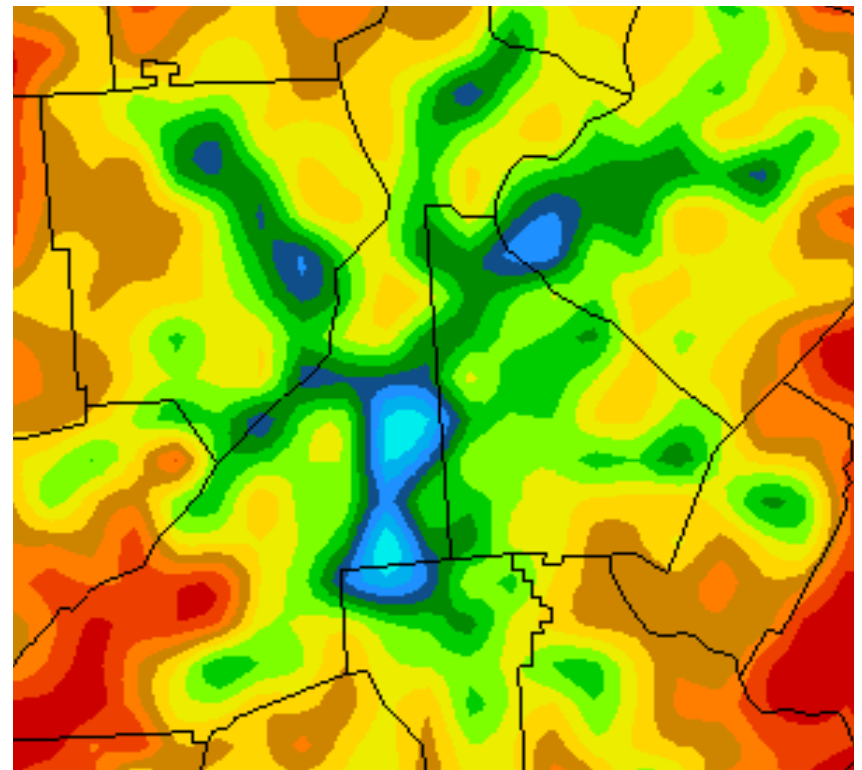
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Vegetation Mitigation Simulation - 2030

2030 Business As Usual



2030 Mitigation Scenario



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-Are native (non-sedum) plants retaining more water than the sedums traditionally used on greenroofs?

-how well do three different sensor methods for measuring water retention compare to each other?

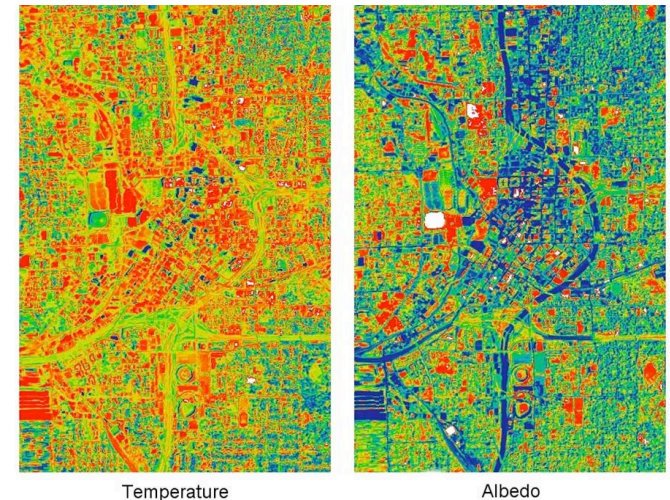


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Hypothesis – Natives/Grasses retain more urban stormwater because they keep stomata open during day unlike sedums ?

Native (NYC) Grasses Planted on Regis HS in NYC



Sedum Plants on Con Ed Facility NYC



We compared the ET from these two projects over the same period of time.

The Native grasses (red) are showing
~32% increase in ET

Monthly Average Latent Heat Flux (Nov 2010 - May 2012)

